

MATH 451/551

Chapter 6. Joint Distribution

6.3 Expected Values

$$\{E(X), E[g(X)]\}$$

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Expected Values



Expected Values

Let X and Y be random variables with joint probability mass function $f(x, y)$ if the random variables are discrete or joint probability density function $f(x, y)$ if the random variables are continuous. The expected value of $g(X, Y)$ is

$$E\{g(X, Y)\} = \begin{cases} \sum_x \sum_y g(x, y) f(x, y) & X, Y \text{ discrete} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f(x, y) dy dx & X, Y \text{ continuous} \end{cases}$$

- The expected value of a function of two random variables $g(X, Y)$ is simply the sum or integral of the product of the functions $g(x, y)$ and $f(x, y)$.

Example 1



Example 1

Find $E(X + Y)$ for the discrete random variables X and Y with joint pmf

$$f(x, y) = \begin{cases} 0.2 & x = 1, y = 1 \\ 0.1 & x = 1, y = 2 \\ 0.3 & x = 1, y = 3 \\ 0.1 & x = 2, y = 1 \\ 0.1 & x = 2, y = 2 \\ 0.2 & x = 2, y = 3 \end{cases}.$$

$$\begin{aligned} E(\underbrace{X+Y}_{g(X,Y)}) &= \sum_A \sum_B g(x, y) f(x, y) \\ &= (1+1) * 0.2 + (1+2) * 0.1 + (1+3) * 0.3 + (2+1) * 0.1 + \\ &\quad (2+2) * 0.1 + (2+3) * 0.2 \\ &= 0.4 + 0.3 + 1.2 + 0.3 + 0.4 + 1.0 = 3.6 \end{aligned}$$

Example 2

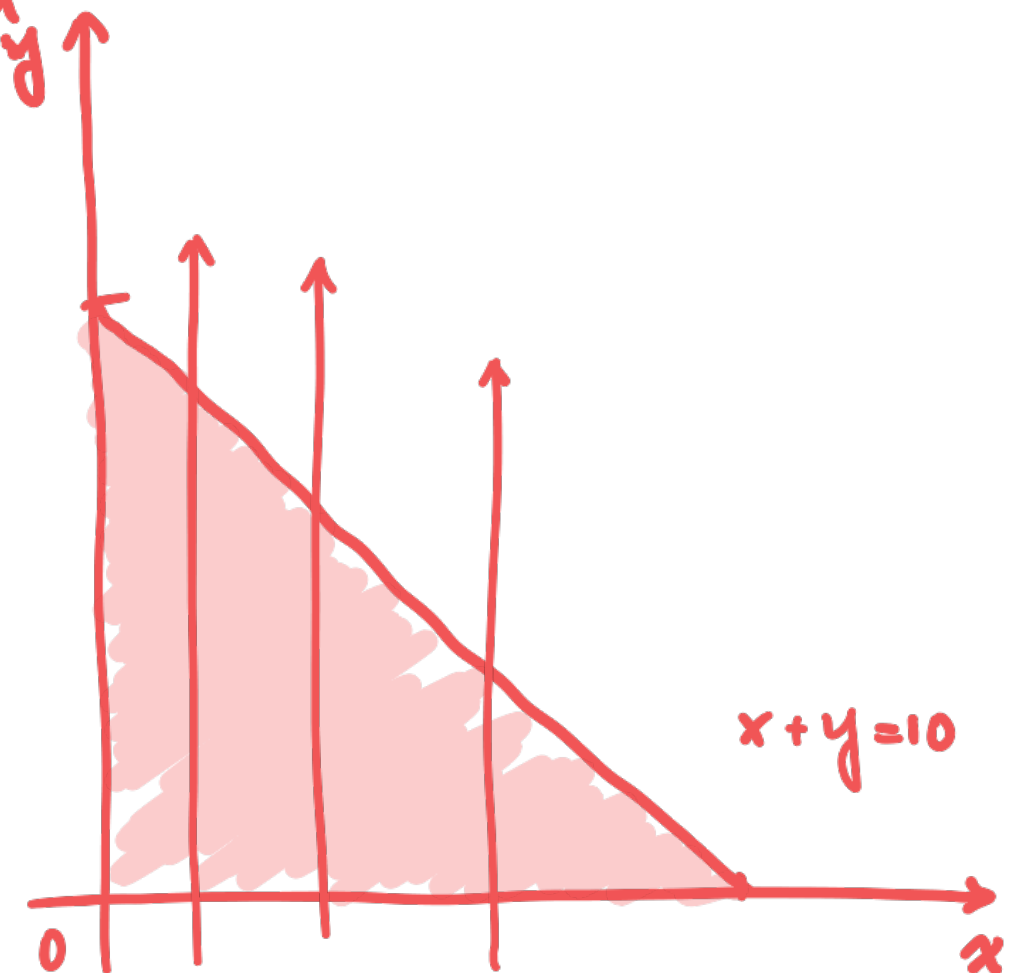


Example 2

Find $E(X^2 Y)$ for X and Y with joint pdf

$$f(x, y) = \frac{1}{50} \quad \underline{x > 0, y > 0, x + y < 10.}$$

$$\begin{aligned} E(\underbrace{X^2 Y}_{g(x, Y)}) &= \int_0^{10} \int_0^{10-x} \underbrace{g(x, y)}_{x^2 y} f(x, y) dy dx \\ &= \int_0^{10} \int_0^{10-x} x^2 y \frac{1}{50} dy dx \\ &= \int_0^{10} \frac{x^2 (10-x)^2}{100} dx \\ &= \frac{100}{3} \end{aligned}$$



Theorem



Theorem 6.2

If X and Y are random variables, then

$$E\{g(X) + h(Y)\} = E\{g(X)\} + E\{h(Y)\}$$

for any functions g and h .

Proof: Let $f(x, y)$ denote the joint probability density fn of X & Y .

$$\begin{aligned} E\{g(X) + h(Y)\} &= \int_{\Omega} \int_{\Omega} \{g(x) + h(y)\} f(x, y) dy dx \\ &= \int_{\Omega} \int_{\Omega} \{ \underbrace{g(x)f(x, y)} + \underbrace{h(y)f(x, y)} \} dy dx \\ &= \int_{\Omega} \int_{\Omega} \underbrace{g(x)f(x, y)} dy dx + \int_{\Omega} \int_{\Omega} \underbrace{h(y)f(x, y)} dy dx \\ &= E\{g(X)\} + E\{h(Y)\} \end{aligned}$$

Theorem



Theorem 6.3

If X and Y are independent random variables, then

$$E\{g(X)h(Y)\} = E\{g(X)\}E\{h(Y)\}$$

for any functions g and h .

Proof. X & Y are independent $\Rightarrow f(x, y) = f_X(x) f_Y(y)$

$$E\{g(X)h(Y)\} = \iint_{\mathcal{A}} g(x)h(y)f(x, y)dydx$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \underbrace{g(x)} \underbrace{h(y)} \underbrace{f_X(x)} \underbrace{f_Y(y)} dy dx$$

$$= \underbrace{\int_{-\infty}^{\infty} g(x)f_X(x)dx} * \int_{-\infty}^{\infty} h(y)f_Y(y)dy$$

$$= E(g(X)) * E\{h(Y)\}$$

Thank You



THANK YOU!